Improving ML training sets with Active Learning in

ML4Astro Catania 2024

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Australian Government Australian Research Council

Acknowledging the traditional owners of Swinburne's land, Wurundjeri People of the Kulin Nation



Rubin LSST

Raw Data Sequential 30s image, 20TB/night



24h

End

Now

Prompt Data Product Difference Image Analysis Alerts: up to 10 million per night



Prompt Products DataBase Images, Object and Source catalogs from DIA Orbit catalog for ~6 million Solar System bodies

Annual Data Release Accessible via the LSST Science Platform &

LSST Data Access Centers.

Final 10yr Data Release Images: 5.5 million x 3.2 Gpx Catalog: 15PB, 37 billion objects

LSST PIs and teams

Publicdatal



Rubin LSST

10 million t-domain candidates/night





broad-science collaboration community driven, open to everyone

Pls & management team Peloton, Ishida, AM



Community driven, join us!



Kilonovae

Orphan, GW+KN, Fast transients



Grandma+Fink 2022,2023 Biswas+2022,2023

Gamma Ray Bursts





Space Awareness

Satellite glints

Karpov+2023,2022

Supernovae SNIa, SN, PISN

Allam+2023, Leoni+2022, Möller+2022 CNRS MITI grant

Nuclear transients

AGN, TDE



Russeil+2022 Gondhalekar+ in prep.

Solar System Objects

Discovery, tracking



Le Montagner+ 2023 Carry+2024 CNRS MITI grant

Anomalies

Pruzhinskaya+ in prep.

LSST simulations: ELAsTiCC

Fraga+ 2024

Type la supernovae



Homogeneous spectral and photometric properties

Direct measurement cosmic expansion

Rubin > 1 million

DES ~ 2k

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(expected 2,000 detected)

Spectroscopic classification



- 1. Supervised + simulations
- 2. Unsupervised
- 3. AL for improving training sets

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SuperNNova

Recurrent Neural Networks

Bayesian Neural Networks (MC dropout & Bayes by Backdrop) *Möller+ 2019, 2022b*

Handles irregular time series

Used with DES, PS, ZTF data Soon Rubin!



Pytorch, many configurations possible

DES: Light-curves + host-galaxy redshifts



Vincenzi, Sullivan, Möller et al. 2021

State-of-the-art simulations

>98% accuracy

Vincenzi+2020



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DES: Light-curves + host-galaxy redshifts





Largest high-z SN Ia sample from a single survey for cosmology

Möller+ 2022a, Vincenzi+ 2024, DES+ 2024

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DES Collaboration 2024



DES Collaboration 2024

DES: Light-curves + host-galaxy redshifts

DES: Light-curves + host-galaxy redshifts



1. Supervised + simulations

2. Unsupervised

3. AL for improving training sets

Unsupervised SN vs AGNs

Preliminary!!!



https://fink-portal.org/download



	Select data source	Data Source	
	Source: ZTF	Choose the type of alerts you want to retrieve	
	Filter alerts Dates: 2021-06-01 - 2022-06-01 Classe(s): ['(SIMBAD) AGN', '(SIMBAD) Blazar', '(SIMBAD) BLLac', '(SIMBAD) LINER', '(TNS) QSO', '(SIMBAD) QSO', '(SIMBAD) Seyfert, '(SIMBAD) Seyfert_1', '(SIMBAD) Seyfert_2', '(TNS) SN', '(TNS) SN I', '(TNS) SN Ia', '(TNS) SN Ia-91bg-like', '(TNS) SN Ia', 91T-like', '(TNS) SN Ia-CSM', '(TNS) SN Ia-pec', '(TNS) SN Iax[02cx-like]', '(TNS) SN Ia', '(TNS) SN Iax[02cx-like]', '(TNS) SN Ia', '(TNS) SN Iax[02cx-like]',	ZTF ELASTICC (v1) ELASTICC (v2.0) ELASTICC (v2.1)	
		Filters Date Range * Pick up start and stop dates (included).	
		June 1, 2021 – June 1, 2022	×
		Alert class Select all classes you like! Default is all classes.	
		(SIMBAD) AGN × (SIMBAD) Blazar × (SIMBAD) BLLac × (SIMBAD) LINER × (TNS) QSO × (SIMBAD) QSO × (SIMBAD) Seyfert ×	
		(SIMBAD) Seyfert_1 × (SIMBAD) Seyfert_2 × (TNS) SN × (TNS) SN I × (TNS) SN Ia × (TNS) SN Ia-91bg-like ×	
SN Ib-pec', '(TNS) SN Ib/c', '(TNS) SN Ibn', '(TNS) SN Ic', '(TNS) SN Ic-BL', '(TNS) SN Ic- pec', '(TNS) SN Icn', '(TNS) SN II', '(TNS) SN	(TNS) SN Ia-91T-like × (TNS) SN Ia-CSM × (TNS) SN Ia-pec × (TNS) SN Iax[02cx-like] × (TNS) SN Ib × (TNS) SN Ib-Ca-rich ×	0	
	(TNS) SN Ib-pec × (TNS) SN Ib/c × (TNS) SN Ibn × (TNS) SN Ic × (TNS) SN Ic-BL × (TNS) SN Ic-pec × (TNS) SN Icn ×		
	II-pec', '(TNS) SN IIb', '(TNS) SN IIL', '(TNS) SN IIn', '(TNS) SN IIn-pec', '(TNS) SN IIP', 'Early SN Ia candidate'] Conditions: candidate.rb >= 0.65; candidate.nbad = 0; candidate.fwhm <= 5; candidate.elong <= 1.2; abs(candidate.magdiff) <= 0.1;	(TNS) SN II × (TNS) SN II-pec × (TNS) SN IIb × (TNS) SN IIL × (TNS) SN IIn × (TNS) SN IIn-pec × (TNS) SN IIP ×	
		(Fink) Early Supernova la candidates ×	
		Extra conditions One condition per line (SQL syntax), ending with semi-colon. See here (and also here) for fields description and here for examples.	
		candidate.rb >= 0.65;	
		candidate.nbad = 0; Recommended filters from the ZTF team	
	Select content Content: Lightcurve	candidate.fwhm <= 5;	
		abs(candidate.magdiff) <= 0.1; to yield a fairly pure sample	
	Submit	Alert content	
	Trigger your job!	Choose the content you want to retrieve	
		Cutouts (~41 KB/alert) Cutouts (~41 KB/alert) Full packet (~55 KB/alert)	
	Description	Submit job	
	_		

Gondhalekar+ in prep.

mTAN: Multi-Time Attention Network



Preliminary!!!



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Preliminary!!!



Gondhalekar+ in prep.

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~0.019 s /light-curve



Compared to the time-series transformer model of Allam, Jr. & McEwen (2021), this is*:

- 1. ~80 times faster inference (0.019s vs. 1.5s).
- 2. ~3 times smaller model size (340 KB vs. 1100 KB).

Assuming ~1 million light curves/year, this amounts to ~0.34 kg of CO_2 , equivalent to driving 1.37 km by an average ICE car^{**}.

*The follow-up paper (Allam, Jr. et al 2023) reduced model size by ~18 and execution time by ~8 times than their original approach.

**Source: MachineLearning Impact calculator

- 1. Supervised + simulations
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3. AL for improving training sets



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Active Learning



Early SNe la

We don't know SNe Ia progenitors Missing modelling for early SNe Ia



Features: sigmoid fit Random Forest

Ishida+2019 Leoni, Ishida+ 2022

Active Learning SN classification (simulated)



Leoni, Ishida et al. 2022

Active Learning SN classification (real data)





2 CV, 16 SN Ia, 5 SN II,1 SN Ib, 2 SN Ibn, 1 SN Ic, 1 microlensing event...

Preliminary!!!



Möller et al. in prep.

Preliminary!!!



Robotic Network @ Siding Spring Observatory

Lidman, Colless, Wolf, Travouillon, Brough, Seitenzahl, Ruiter, Brown, Heger, Galloway, Möller, Kamath, O'Toole, Wen, Einecke



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Automatic data reduction

Rubin is an amazing opportunity and a big data challenge!



- All our data is public <u>fink-portal.org</u>
- First candidate ID modules: SNe, KNe, SSO, GRB, microlensing...

Supernova ML challenges:

- AL is a good strategy to improve training sets (optimise follow-up)
- This will percolate to simulations (built with templates)
- And then we can use large NNs accurately for precision science!





Backup

